

# Statistical Thermodynamics in Chemistry and Biology

## 5. Entropy and the Boltzmann distribution law

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# Agenda

We will continue adding piece by piece to the puzzle...

In this chapter, we will discuss:

- ▶ **Entropy**: connection between entropy, multiplicity, and probability
- ▶ **Boltzmann's distribution law**: a first glance

# What is entropy?

- ▶ Boltzmann's law:

$$S = k_B \ln W$$

- ▶ From a probability perspective, we "know":

$$S = -k_B \sum_{i=1}^t p_i \ln p_i$$

- ▶ The multiplicity  $W$  (e.g. for placing  $t$  types of molecules on a lattice):

$$\begin{aligned} \ln W &= \ln \frac{N!}{n_1! n_2! \dots n_t!} \stackrel{\text{Stirling}}{\approx} N \ln N - N - \sum_{i=1}^t (n_i \ln n_i - n_i) \\ &= - \sum_{i=1}^t n_i \ln \frac{n_i}{N} = -N \sum_{i=1}^t p_i \ln p_i \end{aligned}$$

where we essentially used  $\sum_{i=1}^t n_i = N$  in "both directions".

# What is entropy?

## Part 2

- To continue:

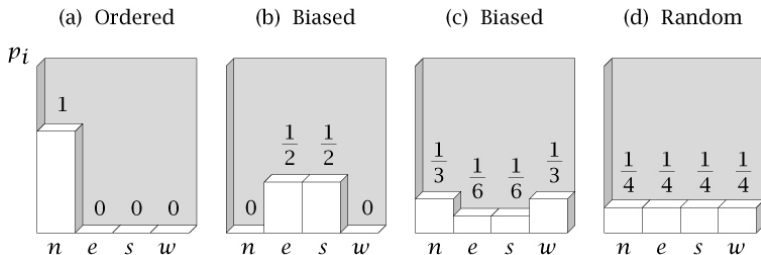
$$\ln W = -N \sum_{i=1}^t p_i \ln p_i \quad \Rightarrow \quad \frac{1}{N} \ln W = - \sum_{i=1}^t p_i \ln p_i = \frac{S_N}{Nk_B}$$

where  $S_N$  is the entropy for  $N$  trials. The normal definition of entropy is per trial (or per particle or per mole)

$$S = \frac{S_N}{N}$$

## Example 5.1: Order and disorder

Spin a pencil  $N$  times: *north*, *east*, *west*, and *south*



$$W = \frac{N!}{n_n!n_e!n_w!n_s!}$$

- ▶ Which case in the figure gives the maximum multiplicity?
- ▶ A water molecule has a dipole moment. Which dipole moment has *liquid water*? *Liquid water has on average a zero dipole moment.*

# Minimization with constraints

See chapter 4, pages 66-72

- ▶ Assume that we would like to minimize a function  $f(x, y)$  with the constraint  $g(x, y) = 0$ .
- ▶ The method of **Lagrange multipliers**: construct a function  $h(x, y)$ ,

$$h(x, y) = f(x, y) - \lambda g(x, y)$$

where  $\lambda$  is a **Lagrange multiplier**.

- ▶ Minimize  $h(x, y)$  with respect to  $x$ ,  $y$  and  $\lambda$ ,

$$\frac{\partial h(x, y)}{\partial x} = 0$$

$$\frac{\partial h(x, y)}{\partial y} = 0$$

$$\frac{\partial h(x, y)}{\partial \lambda} = g(x, y) = 0$$

## Maximum entropy: no (physical) constraints

- ▶ Still, the probability has to sum to 1:

$$\sum_{i=1}^t p_i = 1 \quad \Rightarrow \quad \sum_{i=1}^t dp_i = 0$$

- ▶ We seek the distribution  $(p_1^*, p_2^*, \dots, p_t^*)$  that maximizes the entropy,  $S$ ,

$$S(p_1, p_2, \dots, p_t) = -k_B \sum_{i=1}^t p_i \ln p_i$$

subject to the constraint above.

- ▶ Note the differential of  $S$ :

$$dS = \sum_{i=1}^t \frac{\partial S}{\partial p_i} dp_i$$

# Maximum entropy: no (physical) constraints

## Part 2

- Solve by the Lagrange multiplier method:

$$\sum_{i=1}^t \left( \frac{\partial S}{\partial p_i} - \alpha \right) dp_i = 0$$

where  $\alpha$  is the Lagrange multiplier (including also  $k_B$ ).

- Has to be fulfilled for each  $i$ :

$$-1 - \ln p_i^* - \alpha = 0 \quad \Rightarrow \quad p_i^* = e^{(-1-\alpha)}$$

- To simplify:

$$p_i^* = \frac{p_i^*}{1} = \frac{p_i^*}{\sum_{i=1}^t p_i^*} = \frac{e^{(-1-\alpha)}}{te^{(-1-\alpha)}} = \frac{1}{t}$$

- Conclusion: Even (flat) distribution



## Maximum entropy: constant energy

- ▶ The total energy is constant:

$$E = \sum_{i=1}^t \varepsilon_i n_i \quad \Rightarrow \quad \frac{E}{N} = \sum_{i=1}^t p_i \varepsilon_i$$

- ▶ We seek the distribution that maximizes the entropy with an extra constraint: the probabilities still have to sum to 1 and the energy is  $E/N$  is constant:

$$\sum_{i=1}^t dp_i = 0 ; \quad \sum_{i=1}^t \varepsilon_i dp_i = 0$$

- ▶ Solve by the Lagrange multiplier method:

$$\sum_{i=1}^t \left( \frac{\partial S}{\partial p_i} - \alpha - \beta \varepsilon_i \right) dp_i = 0$$

- ▶ Lagrange multipliers give (analogously to the previous example):

$$-1 - \ln p_i^* - \alpha - \beta \varepsilon_i = 0 \quad \Rightarrow \quad p_i^* = e^{(-1-\alpha-\beta\varepsilon_i)}$$

# Maximum entropy: constant energy

## Part 2

- ▶ Same approach again:

$$p_i^* = \frac{p_i^*}{1} = \frac{p_i^*}{\sum_{i=1}^t p_i^*} = \frac{e^{(-1-\alpha)} e^{-\beta \varepsilon_i}}{\sum_{i=1}^t e^{(-1-\alpha)} e^{-\beta \varepsilon_i}} = \frac{e^{-\beta \varepsilon_i}}{\sum_{i=1}^t e^{-\beta \varepsilon_i}} = \frac{e^{-\beta \varepsilon_i}}{q}$$

- ▶ The Boltzmann distribution law

$$p_i^* = \frac{e^{-\beta \varepsilon_i}}{q}$$

- ▶  $q$  is the partition function

$$q = \sum_{i=1}^t e^{-\beta \varepsilon_i}$$

which is a very central concept in statistical thermodynamics.

- ▶ The Lagrange multiplier,  $\beta$ , is at this stage not determined.

# Summary

- ▶ Defined **entropy** from **Boltzmann's law**.
- ▶ Derived the **Boltzmann distribution law** for two cases.
- ▶ The **partition function** is defined.